

Quantum compilation and its noise resilience

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Slides for [Quantum 3, 140 (2019)] and [NJP 22, 043006 (2020)]

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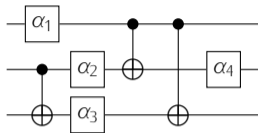
Near-term quantum computing

Ultimate goal: *fault-tolerant quantum computers*. But this requires many good-quality qubits: ≈ 1000 s of physical qubits per logical qubit.

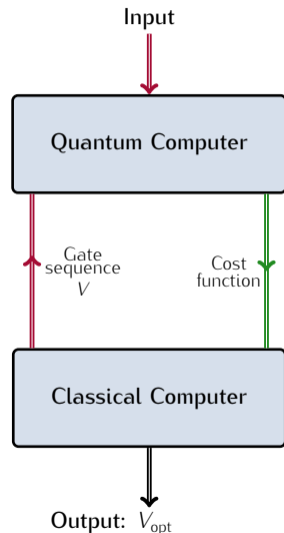
We currently have **noisy intermediate-scale quantum (NISQ)** computers:

- ▶ Limited number of qubits (≈ 50 - 100), not fault tolerant.
- ▶ Limited connectivity between qubits.
- ▶ Limited circuit depth due to noise.

What can we do with such quantum computers?

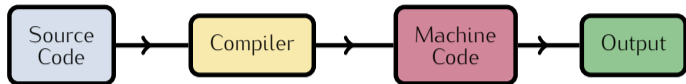


- ▶ Variational quantum eigensolver (VQE).
[Nat. Commun. 5, 4213 (2014)]
- ▶ Variational quantum autoencoders.
[Quant. Sci. Technol. 2, 045001 (2017)]
- ▶ Variational quantum state diagonalization.
[npj Quant. Inf. 5, 57 (2019)]
- ★ **Our variational algorithm:** *Quantum-assisted quantum compiling*
[Quantum 3, 140 (2019)].



Compiling quantum algorithms

To run an algorithm on a quantum computer, we must *compile it*.



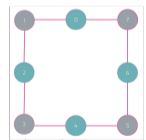
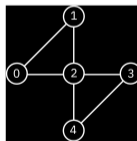
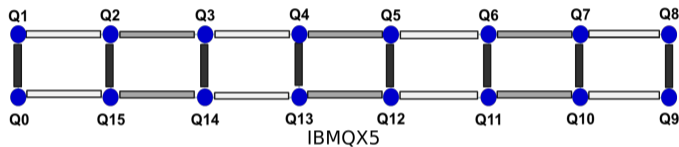
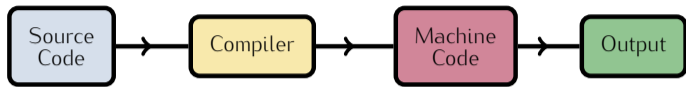
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Need to adapt algorithms to NISQ device constraints:

- ▶ Native gate alphabet;
- ▶ Connectivity;
- ▶ Short depth (use fewest possible gates).

Note: short depth is also important for future fault-tolerant quantum computers.



Compiling quantum algorithms

Typical compilation procedure

[IEEE Trans. CAD 27(3), 436 (2008)]

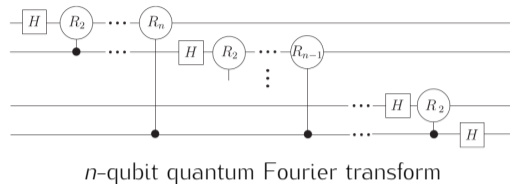
[Quantum Sci. Technol. 3, 025004 (2018)]

Given: Unitary U as a high level gate sequence.

Step 1: Individually decompose every gate into the native gate alphabet.

Step 2: Sweep through the circuit, removing and/or combining redundant gates via, e.g., matrix identities.

Output: Gate sequence V (compilation of U).



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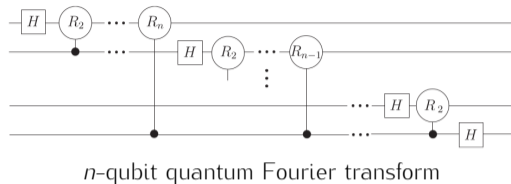
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- ★ Step 2 can be inefficient, esp. for large circuits.
- ★ Requires classical simulation of the circuit in general.

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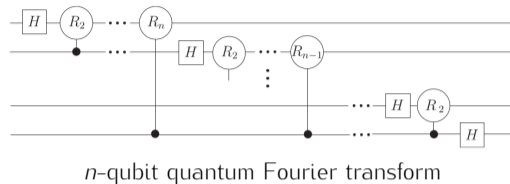
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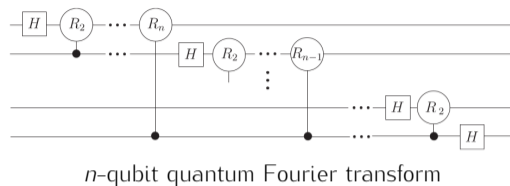
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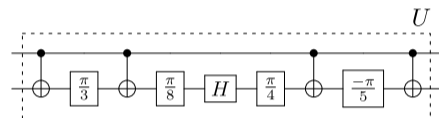


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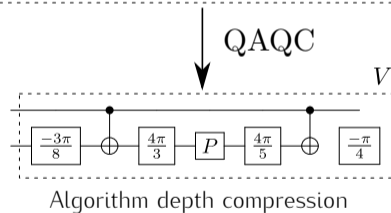
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QAQC performs optimal (approximate) algorithm depth compression

- U is the result of Step 1: an initial “pre-compiled”, non-optimal gate sequence.



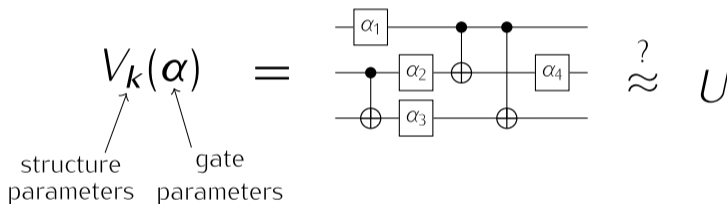
- V is the result of QAQC.



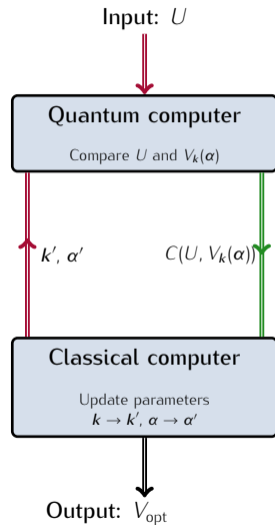
Quantum-assisted quantum compiling (QAQC) [Quantum 3, 140 (2019)]

Given: A unitary U .

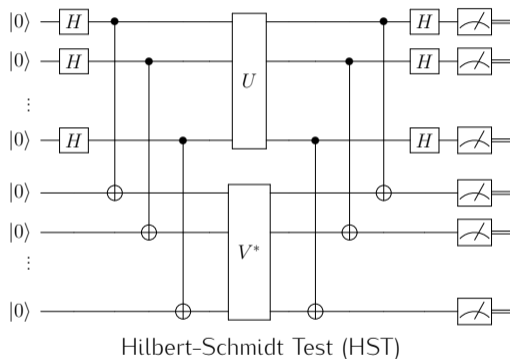
Goal: find the optimal parameterized gate sequence $V_k(\alpha)$.



- ▶ $C(U, V_k(\alpha))$ quantifies the distance between U and $V_k(\alpha)$.
- ▶ $\alpha \rightarrow \alpha'$ via gradient descent optimizer (or non-gradient optimizer);
 $k \rightarrow k'$ via simulated annealing (or genetic algorithms).



Cost function calculation on quantum computer



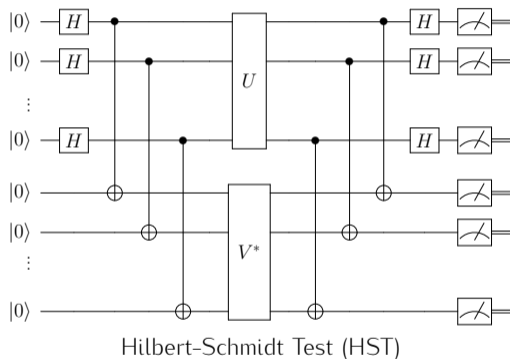
- Probability of the all-zeros outcome is

$$\frac{1}{2^{2n}} |\text{Tr}[V^\dagger U]|^2.$$

The quantity is invariant under global phase.

★ Can also compute gradient (see paper for details).

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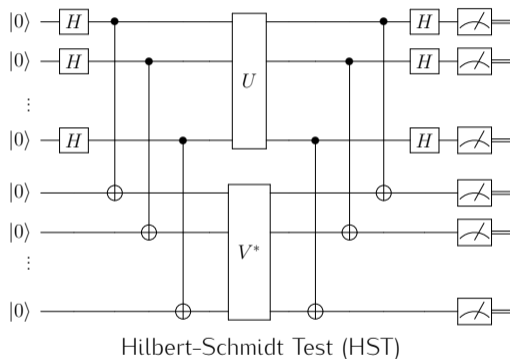
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- Cost function:

$$C_{\text{HST}}(U, V) := 1 - \frac{1}{2^{2n}} |\text{Tr}[V^\dagger U]|^2.$$

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- Related to trace distance:

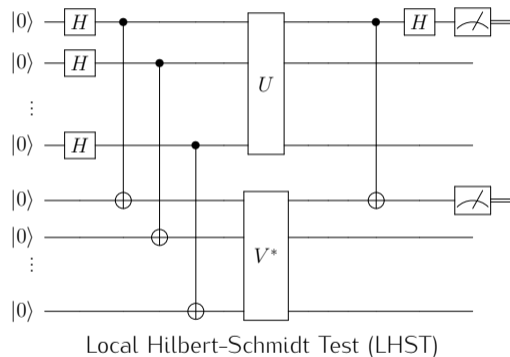
$$\frac{1}{2} \left\| |\Phi_U\rangle\langle\Phi_U| - |\Phi_V\rangle\langle\Phi_V| \right\|_1 = \sqrt{C_{\text{HST}}(U, V)},$$

$$|\Phi_U\rangle = (\mathbb{1} \otimes U)|\Phi\rangle,$$

$$|\Phi_V\rangle = (\mathbb{1} \otimes V)|\Phi\rangle,$$

$|\Phi\rangle$: n -qubit maximally entangled state.

Alternate (local) cost function for large problem sizes



- Cost function:

$$C_{\text{LHST}}(U, V) := \frac{1}{n} \sum_{j=1}^n (1 - \text{Pr}[(0, 0)_j]).$$

- Faithful cost function (like HST).
- Works better for optimization on large problem sizes.

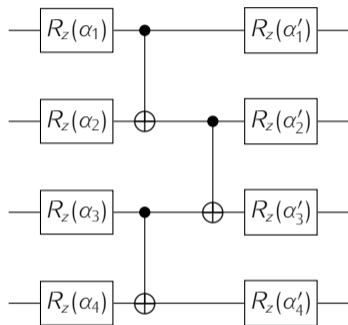
★ Can also compute gradient (see paper for details).

Advantages of QAQC

- ▶ No need to classically simulate the circuit.
 - ▶ Both cost function calculations difficult to simulate classically.
 - ★ Both are DQC1-hard [PRL 81, 5672 (1998)].
 - ★ Classical simulation of DQC1 is impossible unless polynomial hierarchy collapses to the second level [PRL 120, 200502 (2018)].
 - ▶ Holistic approach: optimizes gate parameters and also structure, and does not compile in gate-by-gate manner.
 - ▶ **Not just a NISQ application:** depth compression is useful even for fault-tolerant quantum computers!
- ★ How well does it perform, especially in the presence of noise?

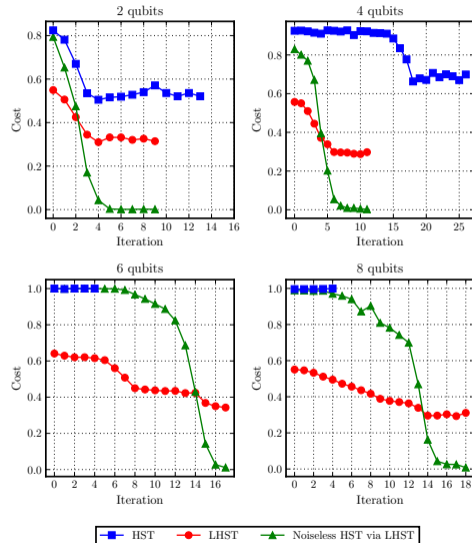
Quantum-assisted quantum compiling (QAQC) [Quantum 3, 140 (2019)]

Results on a noisy simulator



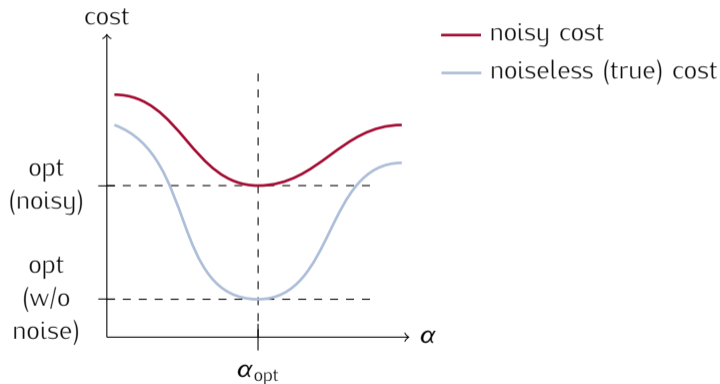
Observations

- ★ HST does not optimize well, but LHST does.
- ★ LHST optimization finds the correct minimum even in the presence of noise!



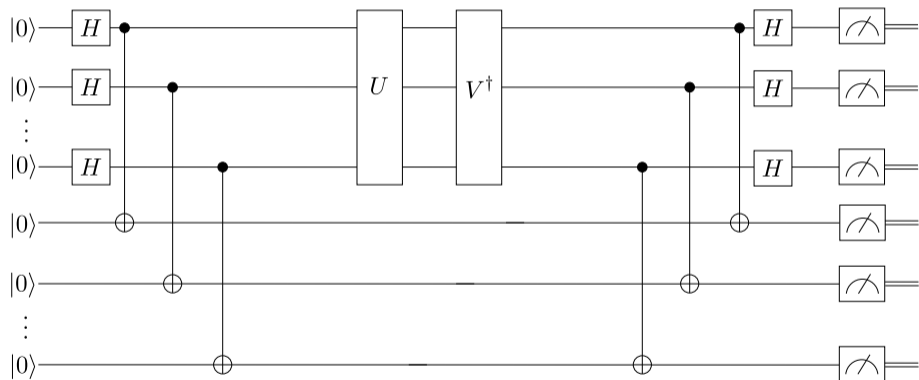
Noise resilience [NJP 22, 043006 (2020)]

- ★ Resilience is with respect to the gate parameters: optimal gate parameters obtained via noisy optimization are also optimal without noise. So we obtain the true optimum even with noise.

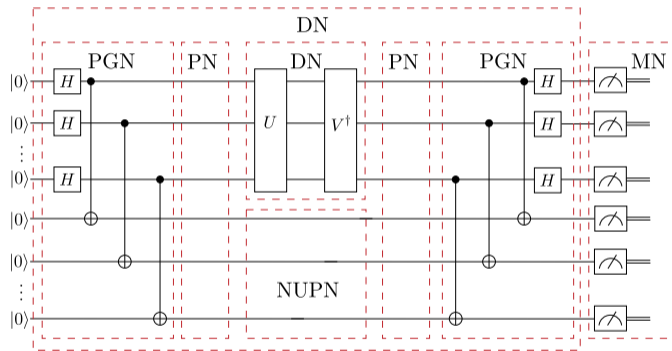


- ★ We provide proofs of noise resilience under certain noise models.

Alternate HST circuit



Noise model 1

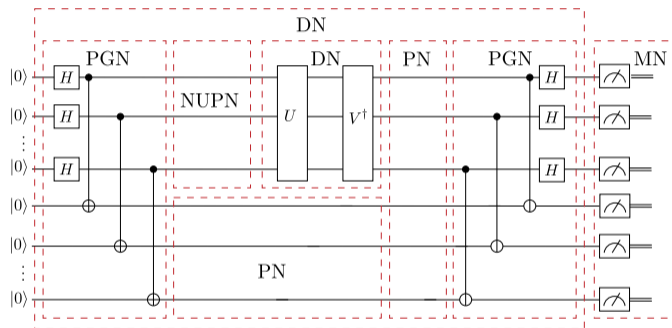


- ▶ DN: depolarizing noise
- ▶ PGN: Pauli gate noise
- ▶ PN: Pauli noise
- ▶ NUPN: non-unital Pauli noise (e.g., amplitude damping)
- ▶ MN: measurement noise

★ We prove noise resilience under this noise model. We can also add noise in between U and V^\dagger (see paper for details).

★ Analogous results for LHST circuit.

Noise model 2

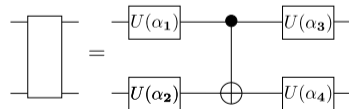
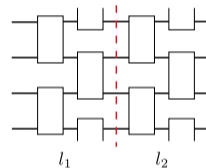
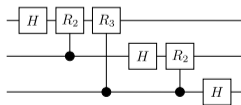
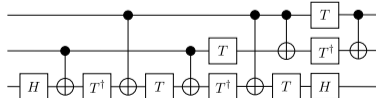
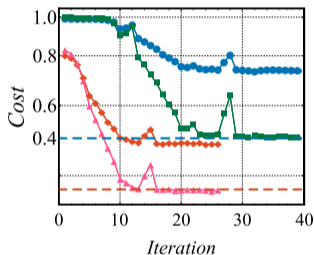
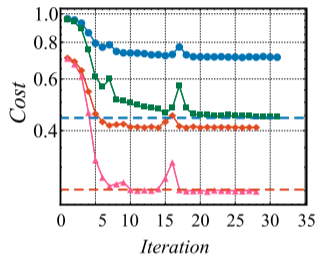


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Examples



- ▶ Variational algorithm (QAQC) to compile quantum algorithms \Rightarrow depth reduction.
[Quantum 3, 140 (2019)].
 - ▶ Circuits (HST and LHST) that compare two quantum circuits.
- ▶ Both the HST and LHST cost functions exhibit *optimal parameter resilience* under certain noise models.
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 - ▶ Optimal solution under noisy optimization corresponds to an optimal solution under noiseless optimization.

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Future directions

- ▶ Investigate noise resilience of other variational algorithms.
- ▶ Algorithms for error mitigation on NISQ devices.

